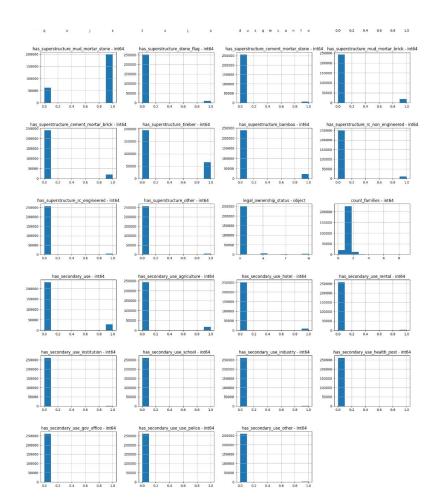
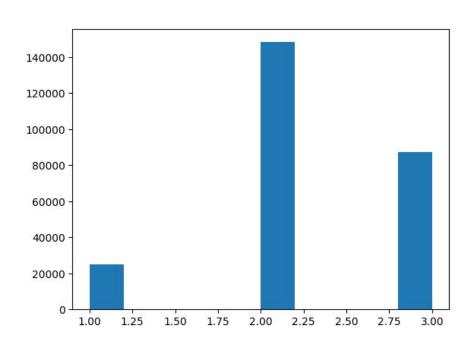
Mini competition

Problem definition





Problem definition



Basic Data exploration

Data cleaning:

Duplicated entries was removed

Outlier removal: using z score

First Basic model

Started with Vanilla Random Forest

- N_estimators: 100

- Max_depth: 6

Performance: $\rightarrow \sim .70$

Modeling improvement

- 1 Random Forest with Gridsearch $\rightarrow \sim .73$
- 2 XGBoost with GridSearch → ~ .728
- 3 Stacking with:

GaussianNB
RandomForestClassifier
DecisionTreeClassifier
AdaBoostClassifier
LinearDiscriminantAnalysis
GradientBoostingClassifier
LogisticRegression
KNeighborsClassifier
LGBMClassifier
ExtraTreesClassifier $\rightarrow \sim .7325$

The following explorations were focussed on improving Stacking

Feature Engineering

```
categorical feature
'geo level 1 id',
'geo level 2 id',
'geo level 3 id',
'land surface condition',
'foundation type',
'roof type',
'ground floor type',
'other floor type',
'position'.
'plan configuration',
'has superstructure adobe mud',
'has superstructure mud mortar stone',
'has superstructure stone flag',
'has superstructure cement mortar stone',
'has superstructure mud mortar brick',
'has superstructure cement mortar brick',
'has superstructure timber',
'has superstructure bamboo',
'has superstructure rc non engineered',
'has superstructure rc engineered',
'has superstructure other',
'legal ownership status',
'has secondary use',
'has secondary use agriculture',
'has secondary use hotel',
```

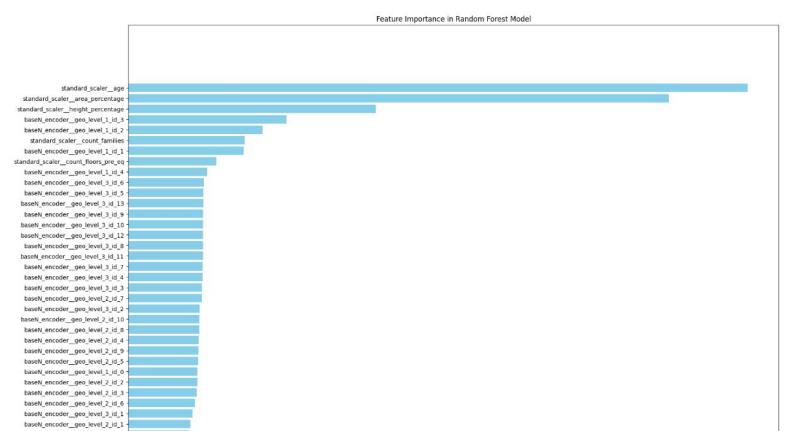
```
df["sticking material"] = df["sticking material"].map({
     'has superstructure mud mortar stone': 'mud',
     'has superstructure mud mortar brick': 'mud',
     'has superstructure cement mortar stone': 'cement',
     'has superstructure cement mortar brick': 'cement'
 }).fillna("none")
df["building material"] = df["building material"].map({
    'has superstructure adobe mud': 'adobe',
    'has superstructure mud mortar stone': 'stone',
    'has superstructure stone flag': 'stone',
    'has superstructure mud mortar brick': 'brick',
    'has superstructure cement mortar stone': 'stone',
    'has superstructure cement mortar brick': 'brick',
    'has superstructure timber': 'wood',
    'has superstructure bamboo': 'wood'
  .fillna("other")
```

Feature engineering

```
categorical feature
'geo level 1 id',
'geo level 2 id',
'geo level 3 id',
'land surface condition',
'foundation type',
'roof type',
'ground floor type',
'other floor type',
'position'.
'plan configurat<u>ion',</u>
'has superstructure adobe mud',
'has superstructure mud mortar stone',
'has superstructure stone flag',
'has superstructure cement mortar stone',
'has superstructure mud mortar brick',
'has superstructure cement mortar brick',
'has superstructure timber',
'has superstructure bamboo',
'has superstructure rc non engineered',
'has superstructure rc engineered',
'has superstructure other',
'legal ownership status',
'has secondary use'.
'has secondary use agriculture',
'has secondary use hotel',
```

```
df["type of building"] = df["type of building"].map({
    'has secondary use agriculture': "agriculture",
    'has secondary use hotel': "institutional",
    'has secondary use rental': "other",
    'has secondary use institution': "institutional",
    'has secondary use school': "institutional",
    'has secondary use industry': "industrial",
    'has secondary use health post': "other",
    'has secondary use gov office': "institutional",
    'has secondary use use police': "institutional",
    'has secondary use other': "other",
    'has secondary use': "other"
}).fillna("other")
```

Feature importance



Feature engineering

```
# Experiment :

# with different types of features :

# 1. all features (all old features )

# 2. all features + new build Features

# 3. all features + new build Features - has_flags

# 4 . all features - has_flags
```

- 0.7201188618338072
- 2 0.7247151377813883
- 3 0.7247151377813883
- 4 0.7201188618338072

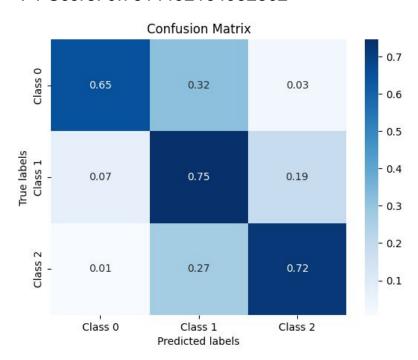
Encoding and scaling

F	1	1 1 2 1 1	11-2-1-1
Experiment ID	geo_level_1_id	geo_level_2_id	geo_level_3_id
Exp-1	Target	Target	Target ~
Exp-2	BaseN	BaseN	BaseN
Exp-3	Frequency	Frequency	Frequency
Exp-4	BaseN	Target	Target
Exp-5	Target	BaseN	BaseN
Exp-6	Frequency	BaseN	BaseN
Exp-7	BaseN	Frequency	Target
Exp-8	Frequency	Target	BaseN

```
Testing Encoding: {'geo_level_1_id': 'target', 'geo_level_2_id': 'target', 'geo_level_3_id': 'target'}
F1 Score: 0.7283
Testing Encoding: {'geo_level_1_id': 'basen', 'geo_level_2_id': 'basen', 'geo_level_3_id': 'basen'}
F1 Score: 0.7216
Testing Encoding: {'geo_level_1_id': 'frequency', 'geo_level_2_id': 'frequency', 'geo_level_3_id': 'frequency'}
F1 Score: 0.7092
Testing Encoding: {'geo_level_1_id': 'basen', 'geo_level_2_id': 'target', 'geo_level_3_id': 'target'}
F1 Score: 0.7276
Testing Encoding: {'geo_level_1_id': 'target', 'geo_level_2_id': 'basen', 'geo_level_3_id': 'basen'}
F1 Score: 0.7221
Testing Encoding: {'geo_level_1_id': 'frequency', 'geo_level_2_id': 'basen', 'geo_level_3_id': 'basen'}
F1 Score: 0.7191
Testing Encoding: {'geo_level_1_id': 'basen', 'geo_level_2_id': 'frequency', 'geo_level_3_id': 'target'}
F1 Score: 0.7265
Testing Encoding: {'geo_level_1_id': 'frequency', 'geo_level_2_id': 'target', 'geo_level_3_id': 'basen'}
F1 Score: 0.7273
```

Encoding

F1 Score: 0.7314492164952862



Modelling

Added new models to Stacking:

```
HistGradientBoostingClassifier
KNeighborsClassifier
RidgeClassifier
QuadraticDiscriminantAnalysis
SGDClassifier
```

KNClass is repeated but with different numbers of neighbours (15)

- Increased the size of models parameters (e.g. n_estimators, iterations...)
- Added log transformed numerical variables to the features

Final Result

Submissions

- To help you track your progress during the competition, each submission is scored against publically available test data to give a "public score".
- The primary evaluation metric is Micro-averaged F1 score. Show more.

